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Robustness of the influence of El Niño on the spatial extent of tropical drought

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Abstract. It has been well established that drought frequently emerges in several tropical regions following the onset of El Niño. An important characteristic of such droughts is their spatial extent, which has recently been linked to the strength of a particular El Niño event. Here the robustness of this relationship is examined through comparison of results from several gridded precipitation analyses for the global tropics. It is found that there is very good agreement across these datasets, including those which incorporate satellite rainfall estimates, confirming that maximum drought extent in tropical land areas increases nearly linearly with associated maximum sea surface temperature anomalies in the east-central tropical Pacific during El Niño events.

1 Introduction

Observationally-based analyses have clearly established that the development of drought in many tropical land areas around the globe is one manifestation of the El Niño phenomenon (e.g., Ropelewski and Halpert, 1987; Mason and Goddard, 2001). Such droughts, sometimes severe, can have serious consequences for human populations within the affected regions. It is therefore of both scientific and practical interest to examine how large an influence El Niño has on the development of tropical drought. Previous studies by Dai et al. (1997, 1998) have shown, based on global, gridded analyses of the Palmer Drought Severity Index (PDSI), that the leading mode of global variability is related to ENSO. Using a standardized precipitation anomaly index based on monthly precipitation and focusing on tropical land areas only, Lyon (2004) recently reported that the spatial extent of drought is closely tied to the strength of an individual El Niño event. The “strength” of an event was defined as the maximum sea surface temperature (SST) anomaly averaged across the Niño 3.4 region (120–170 W, 5 S–5 N) observed during the life cycle

of a given El Niño. Since the spatial distribution of rainfall observing stations across tropical land areas is irregular, gridded rainfall analyses based on these observations are a less than a perfect representation of actual rainfall. Here the robustness of this latter study is examined by comparing results obtained from a number of different rainfall analyses, including those which incorporate satellite estimates of rainfall to supplement the sometimes sparse land-based station coverage.

2 Data and methodology

2.1 Data

Two basic types of gridded rainfall analyses were employed: those based solely on observed monthly precipitation at land-based recording stations; and merged analyses which incorporate both station data and satellite estimates of monthly rainfall. Analyses of the first type include the precipitation reconstruction over land (PRECL; Chen et al., 2002) dataset on a $2.5^\circ \times 2.5^\circ$ latitude/longitude grid; two analyses from the Climate Research Unit at the University of East Anglia (New et al., 2000; Hulme, 1994), at $0.5^\circ \times 0.5^\circ$ and $2.5^\circ \times 3.75^\circ$ latitude/longitude resolution respectively; and the Climate Anomalies Monitoring System (CAMS; Ropelewski et al., 1984) dataset from the US Climate Prediction Center (CPC) on a $2.0^\circ \times 2.0^\circ$ latitude/longitude grid. Of the second type are the Merged Analysis of Precipitation from the CPC (CMAP; Xie and Arkin, 1996) and the merged analysis from the Global Precipitation Climatology Project (GPCP; Huffman et al., 1997), both on a $2.5^\circ \times 2.5^\circ$ latitude/longitude grid. The spatial domain for all datasets used is 30°S – 30°N , land areas only. The base period for computing monthly anomalies is 1961–1990 for the land-based datasets, and 1979–2003 for the merged analyses. The period of analysis is from 1950–1998 for the land-based analyses, and 1979–2003 for the merged analyses.

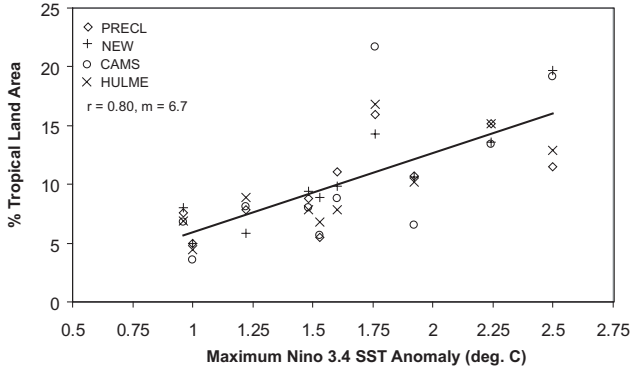


Fig. 1. Maximum percentage of tropical land area (excluding desert regions) experiencing intermediate drought (<-1.5 index value) as a function of associated maximum SST anomaly ($^{\circ}\text{C}$) for the Nino 3.4 region during the 10 strongest El Niño events during 1950–1998. Symbols in upper-left indicate the various rainfall analyses used.

The SST data used was from Smith and Reynolds (2003) with a base period of 1961–1990 used for computing monthly anomalies. A 5-month running average was applied to SST anomalies averaged across the Nino 3.4 region. An El Niño event was defined as occurring when the Nino 3.4 SST anomaly exceeded 0.5°C for at least six consecutive months.

2.2 Defining drought and measuring its spatial coverage

Using the methodology of Lyon and Barnston (2005) drought events were defined based on 12-month overlapping sums of weighted, standardized precipitation anomalies,

$$S_{12} = \sum_{i=1}^{12} \left(\frac{P_i - \bar{P}_i}{\sigma_i} \right) \cdot \frac{\bar{P}_i}{P_A} \quad (1)$$

In Eq. (1), P_i is the monthly precipitation of the i th month in the sum at a given gridpoint, overbars represent climatological means, and σ_i is the standard deviation of the monthly precipitation. P_A is the total annual precipitation at a gridpoint, and the weighting factor, P_i/P_A , dampens standardized anomalies during climatological dry seasons while emphasizing those during rainy seasons. The sum S_{12} is then itself standardized to obtain a dimensionless indicator of the relative severity of drought (when the index is negative). As in Lyon (2004), three levels of drought severity were considered based on the standardized index: *moderate* (<-1.0), *intermediate* (<-1.5), and *severe* (<-2.0). Climatologically dry regions (deserts) were masked from the analysis.

To determine the spatial extent of drought, the three different levels of drought severity were considered separately for each dataset. Gridpoints with standardized rainfall index values below a given threshold (i.e., -1.0 , -1.5 , -2.0) were flagged, weighted by the cosine of their latitude, and the fraction of all flagged gridpoints within tropical land areas then computed (expressed as percentages). From these time series, the maximum percentage of land area in drought during

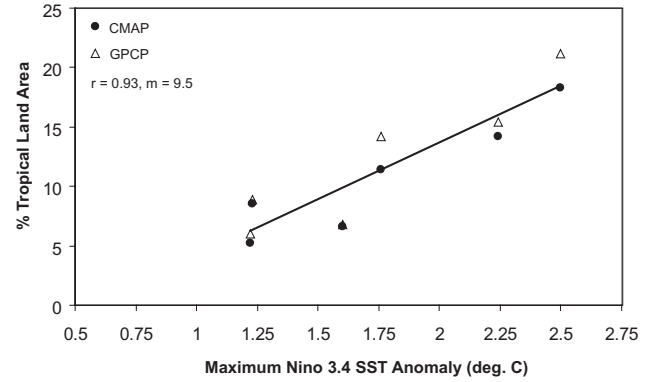


Fig. 2. As in Fig. 1, but based on data from two merged precipitation analyses covering the period 1979–2003.

a particular El Niño event was obtained and related to the associated maximum in SST anomaly in the Nino 3.4 region.

3 Results

The 10 strongest El Niño events during 1950–1998 were used to produce Fig. 1 which displays the maximum spatial extent of *intermediate* drought (standardized rainfall index values <-1.5) as a function of maximum Nino 3.4 SST anomaly. Analyses of other levels of severity produced similar results. The solid line on the plot is based on a linear regression of the mean values of peak spatial extent measured across the datasets for each El Niño event. The positive slope of the regression line indicates that the spatial extent of drought across the tropics increases as Nino 3.4 SST anomalies (the strength of El Niño events) increase. The linear correlation coefficient is 0.8. Consistent with previous results (based on the New dataset), the spatial extent of drought increases by roughly a factor of two between the weakest and strongest events. The greatest departure from the regression line occurs during the 1991–92 El Niño (Nino 3.4 SST anomaly of $\approx 1.75^{\circ}\text{C}$), particularly for the CAMS dataset, with the other analyses also indicating increased coverage relative to the linear fit.

Figure 2 is the same analysis, but based on the CMAP and GPCP merged precipitation analyses. The two datasets are seen to be in very good agreement, with a linear correlation for the average extent between the two exceeding 0.9. Also noteworthy is that the spatial extent of drought during the 1991–92 El Niño (again, with associated Nino 3.4 SST anomaly of $\approx 1.75^{\circ}\text{C}$) is in much closer agreement with the linear fit to the data than for the purely land-based analyses shown in Fig. 1.

4 Conclusions

Six different gridded rainfall analyses covering tropical land areas within 30° latitude of the equator were utilized to test the robustness of the relationship between the strength of

El Niño events and the peak spatial extent of associated droughts. Each of the datasets indicated an increase in spatial extent of drought with increasing strength of El Niño, confirming earlier results. For all of the datasets, drought extent was found to approximately double between the weakest and strongest El Niño events which have occurred over the past half-century. There is some evidence that satellite estimates of rainfall in the tropics may help in realistically capturing actual amounts, at least to the extent to which the linear regression analyses presented are a true representation of drought characteristics in the tropics.

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